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```
library(e1071) library(caret)
data=read.csv("C:/Users/91828/Downloads/Cleaned-Data.csv",
header=TRUE)
# taking first 1000 rows
data=data[0:1000,]
# remove Na values
data=na.omit(data)
str(data)
                   1000 obs. of 27 variables:
## 'data.frame':
## $ Fever : int 1 1 1 1 1 1 1 1 1 1 ... ## $ Tiredness
int 1 1 1 1 1 1 1 1 1 1 ...
## $ Dry.Cough
                : int 1 1 1 1 1 1 1 1 1 1 ... ## $
Difficulty.in.Breathing: int 1 1 1 1 1 1 1 1 1 1 ... ## $
Sore.Throat : int 1 1 1 1 1 1 1 1 1 1 ... ## $ None Sympton :
int 0 0 0 0 0 0 0 0 0 0 ... ## $ Pains : int 1 1 1 1 1 1 1
1 1 1 ... ## $ Nasal.Congestion
                               : int 1 1 1 1 1 1 1 1 1 1
                      : int 1 1 1 1 1 1 1 1 1 1 ... ## $
... ## $ Runny.Nose
Diarrhea
          : int 1 1 1 1 1 1 1 1 1 1 ... ## $
None Experiencing : int 0 0 0 0 0 0 0 0 0 ... ## $ Age 0.9
      : int 1 1 1 1 1 1 1 1 1 1 ... ## $ Age 10.19
0000000000...
## $ Age 20.24
                            : int 0 0 0 0 0 0 0 0 0 0 ...
## $ Age 25.59 : int 0 0 0 0 0 0 0 0 0 ... ## $ Age 60.
      : int 0 0 0 0 0 0 0 0 0 ... ## $ Gender Female
int 0 0 0 0 0 0 0 0 0 0 ... ## $ Gender Male
                                            : int 1 1 1 1
1 1 1 1 1 1 ... ## $ Gender Transgender : int 0 0 0 0 0 0
000 ... ## $ Severity Mild : int 1 1 1 0 0 0 0 0 0 0 ...
## $ Severity_Moderate : int 0 0 0 1 1 1 0 0 0 0 ... ## $
Severity None
                : int 0 0 0 0 0 0 0 0 1 ... ## $
Severity Severe : int 0 0 0 0 0 0 1 1 1 0 ... ## $
Contact_Dont.Know : int 0 0 1 0 0 1 0 0 1 0 ... ## $
Contact No : int 0 1 0 0 1 0 0 1 0 0 ... ## $ Contact Yes :
int 1 0 0 1 0 0 1 0 0 1 ...
                             : chr "China" "China" "China" ...
## $ Country
attach(data)
set.seed(101)
# Logistic Regression
```

```
## For Mild Severity
fit=glm(Severity_Mild~Dry.Cough+Difficulty.in.Breathing,data=data,famil
y=binomial) summary(fit)
##
## Call:
## glm(formula = Severity_Mild ~ Dry.Cough + Difficulty.in.Breathing,
       family = binomial, data = data)
##
##
## Deviance Residuals:
       Min
                  10 Median
                                   3Q
                                           Max
## -0.7753 -0.7585 -0.7585 1.6423 1.6651
##
## Coefficients:
##
                           Estimate Std. Error z value Pr(>|z|)
## (Intercept)
                          -1.09861 0.14213 -7.729 1.08e-14 ***
## Dry.Cough
                            0.05064
                                        0.21264 0.238
                                                          0.812
## Difficulty.in.Breathing -0.05064 0.18738 -0.270 0.787 ## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## (Dispersion parameter for binomial family taken to be 1)
##
        Null deviance: 1129 on 999 degrees of freedom
## Residual deviance: 1129 on 997 degrees of freedom
## AIC: 1135
## Number of Fisher Scoring iterations: 4
# making predictions using logistic regression model
probab=predict(fit,type="response")
pred=ifelse(probab>0.5,1,0)
table(pred, Severity_Mild)
##
       Severity Mild
## pred 0 1 ##
     0 748 252
mean(pred==Severity Mild)
## [1] 0.748
## training subset
train=(Difficulty.in.Breathing==0)
trainingset=data[!train,]
trainingset_severityMmild=Severity_Mild[train]
## fit 2 using training set
fit2=glm(Severity_Mild~Dry.Cough+Difficulty.in.Breathing,data=data,fami
```

```
ly=binomial, subset=train)
probab2=predict(fit2, type="response")
pred2=ifelse(probab2>0.5,1,0)
table(pred2,trainingset_severityMmild)
       trainingset_severityMmild
## pred2 0 1 ##
     0 352 120
mean(pred2==trainingset_severityMmild)
## [1] 0.7457627
#loading data again
data2=read.csv("C:/Users/ishiv/Downloads/Cleaned-Data.csv", header=TRUE
# taking first 1000 rows
data2=data2[0:1000,]
# remove Na values
data2=na.omit(data2)
str(data2)
## 'data.frame': 1000 obs. of 27 variables:
## $ Fever : int 1 1 1 1 1 1 1 1 1 1 ... ## $ Tiredness
int 1 1 1 1 1 1 1 1 1 1 ...
## $ Dry.Cough
                : int 1 1 1 1 1 1 1 1 1 1 ... ## $
Difficulty.in.Breathing: int 1 1 1 1 1 1 1 1 1 1 ... ## $
Sore.Throat : int 1 1 1 1 1 1 1 1 1 1 ... ## $ None Sympton :
int 0 0 0 0 0 0 0 0 0 0 ... ## $ Pains : int 1 1 1 1 1 1
1 1 1 ... ## $ Nasal.Congestion : int 1 1 1 1 1 1 1 1 1
... ## $ Runny.Nose
                    : int 1 1 1 1 1 1 1 1 1 1 ... ## $
           : int 1 1 1 1 1 1 1 1 1 1 ... ## $
Diarrhea
None Experiencing : int 0 0 0 0 0 0 0 0 0 ... ## $ Age 0.9
     : int 1 1 1 1 1 1 1 1 1 1 ... ## $ Age 10.19
000000000...
## $ Age 20.24
                            : int 0 0 0 0 0 0 0 0 0 0 ...
## $ Age_25.59 : int 0 0 0 0 0 0 0 0 0 0 ... ## $ Age_60.
     : int 0 0 0 0 0 0 0 0 0 0 ... ## $ Gender Female
int 0 0 0 0 0 0 0 0 0 0 ... ## $ Gender Male
                                            : int 1 1 1 1
1 1 1 1 1 1 ... ## $ Gender_Transgender : int 0 0 0 0 0 0 0
0 0 0 ... ## $ Severity Mild : int 1 1 1 0 0 0 0 0 0 0 ...
## $ Severity_Moderate : int 0 0 0 1 1 1 0 0 0 0 ... ## $
Severity None
               : int 0 0 0 0 0 0 0 0 1 ... ## $
Severity Severe : int 0 0 0 0 0 0 1 1 1 0 ... ## $
Contact Dont.Know : int 0 0 1 0 0 1 0 0 1 0 ... ## $
Contact_No : int 0 1 0 0 1 0 0 1 0 0 ...
                                                                    )
```

```
: int 1 0 0 1 0 0 1 0 0 1 ...
## $ Contact Yes
## $ Country
                              : chr "China" "China" "China" ...
## SVM
svm_train <- sample(c(TRUE, FALSE), nrow(data2), replace = T, prob = c</pre>
(0.7, 0.3)
# splitting in test and train set
train2 <- data2[svm train, ]</pre>
testing <- data2[!svm train, ]</pre>
x=train2[c('Dry.Cough','Difficulty.in.Breathing')]
y=train2$Severity Mild
data2=data.frame(x=x,y=as.factor(y))
# using linear kernel
tune.out=tune(svm, y ~., data=data2 ,kernel='linear', ranges = list(cos
t=c(0.001,0.1,1,10,100))) summary(tune.out)
##
## Parameter tuning of 'svm':
## - sampling method: 10-fold cross validation
##
## - best parameters:
## cost
## 0.001
##
## - best performance: 0.2383216
##
## - Detailed performance results:
      cost
               error dispersion
## 1 1e-03 0.2383216 0.05323898
## 2 1e-01 0.2383216 0.05323898
## 3 1e+00 0.2383216 0.05323898
## 4 1e+01 0.2383216 0.05323898
## 5 1e+02 0.2383216 0.05323898
# getting the best model
bestmodel=tune.out$best.model
summary(bestmodel)
##
## Call:
## best.tune(method = svm, train.x = y \sim ., data = data2, ranges = list
(cost = c(0.001,
##
       0.1, 1, 10, 100)), kernel = "linear")
##
```

##		

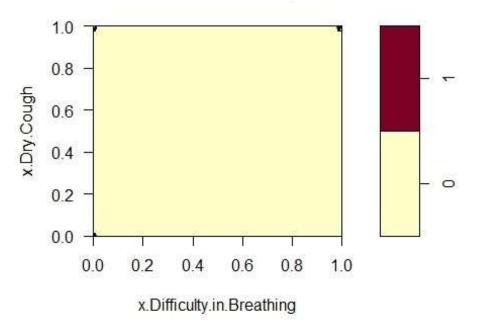
```
## Parameters:
##
      SVM-Type: C-classification
## SVM-Kernel: linear
##
          cost: 0.001
##
## Number of Support Vectors: 340
##
## ( 170 170 )
##
##
## Number of Classes: 2
##
## Levels:
## 0 1
# plot
plot(bestmodel, data2)
xt=testing[c('Dry.Cough','Difficulty.in.Breathing')]
yt=testing$Severity_Mild
tdata=data.frame(x=xt,y=as.factor(yt))
ypred=predict(bestmodel,tdata)
confusionMatrix(ypred,tdata$y)
## Confusion Matrix and Statistics
##
             Reference
##
## Prediction 0 1
##
            0 205 82
              100
##
##
##
                  Accuracy : 0.7143
                    95% CI: (0.6582, 0.7658)
##
       No Information Rate: 0.7143
##
##
       P-Value [Acc > NIR] : 0.5297
##
##
                     Kappa: 0
## Mcnemar's Test P-Value : <2e-16
##
##
               Sensitivity: 1.0000
##
               Specificity: 0.0000
##
      Pos Pred Value : 0.7143 ##
     Neg Pred Value : NaN
##
                Prevalence: 0.7143
            Detection Rate: 0.7143
##
##
      Detection Prevalence: 1.0000
##
         Balanced Accuracy: 0.5000
```

## ## 'Positive' Class : 0 ##

```
#radial kernel
tune.out.r1=tune(svm, y ~., data=data2 ,kernel='radial', ranges = list
(cost=c(0.001,0.1,1,10,100)),gamma=1)
summary(tune.out.r1)
##
## Parameter tuning of 'svm':
## - sampling method: 10-fold cross validation
## - best parameters:
## cost
## 0.001
## - best performance: 0.2383803
##
## - Detailed performance results:
      cost error dispersion ## 1
1e-03 0.2383803 0.0497398
## 2 1e-01 0.2383803 0.0497398
## 3 1e+00 0.2383803 0.0497398
## 4 1e+01 0.2383803 0.0497398 ##
5 1e+02 0.2383803 0.0497398
bestmodel.r1=tune.out.r1$best.model
summary(bestmodel.r1)
##
## Call:
## best.tune(method = svm, train.x = y \sim ., data = data2, ranges = list
(cost = c(0.001,
       0.1, 1, 10, 100), kernel = "radial", gamma = 1)
##
##
##
## Parameters:
      SVM-Type: C-classification
## SVM-Kernel: radial
##
          cost: 0.001
##
## Number of Support Vectors: 340
##
## ( 170 170 )
##
##
## Number of Classes: 2
## Levels:
```

## 0 1
plot(bestmodel.r1,data2)

## SVM classification plot



```
xtb=testing[c('Dry.Cough','Difficulty.in.Breathing')]
ytb=testing$Severity_Mild
tdata=data.frame(x=xtb,y=as.factor(ytb))
ypred.r=predict(bestmodel.r1,tdata)
confusionMatrix(ypred.r,tdata$y)
## Confusion Matrix and Statistics
##
##
             Reference
## Prediction 0 1
            0 205 82
##
              100
##
##
##
                  Accuracy : 0.7143
##
                    95% CI: (0.6582, 0.7658)
##
       No Information Rate: 0.7143
       P-Value [Acc > NIR] : 0.5297
##
##
##
                     Kappa: 0
##
## Mcnemar's Test P-Value : <2e-16
##
##
               Sensitivity: 1.0000
##
               Specificity: 0.0000
```

```
## Pos Pred Value : 0.7143 ##
Neg Pred Value : NaN
## Prevalence : 0.7143

## Detection Rate : 0.7143
```

##

##

## ## ## Detection Prevalence : 1.0000 Balanced Accuracy : 0.5000

'Positive' Class : 0